

23MA301: MATHEMATICAL STATISTICAL FOUNDATION

Topic: RANDOM VARIABLES

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Concept of a Random Variable

A **random variable** assigns a numerical value to each outcome of a random experiment. Represented by it maps uncertain outcomes to numbers — depending entirely on chance.

- **Discrete:** Countable outcomes — e.g., rolling a die, coin toss
- **Continuous:** Measurable values — e.g., height, temperature

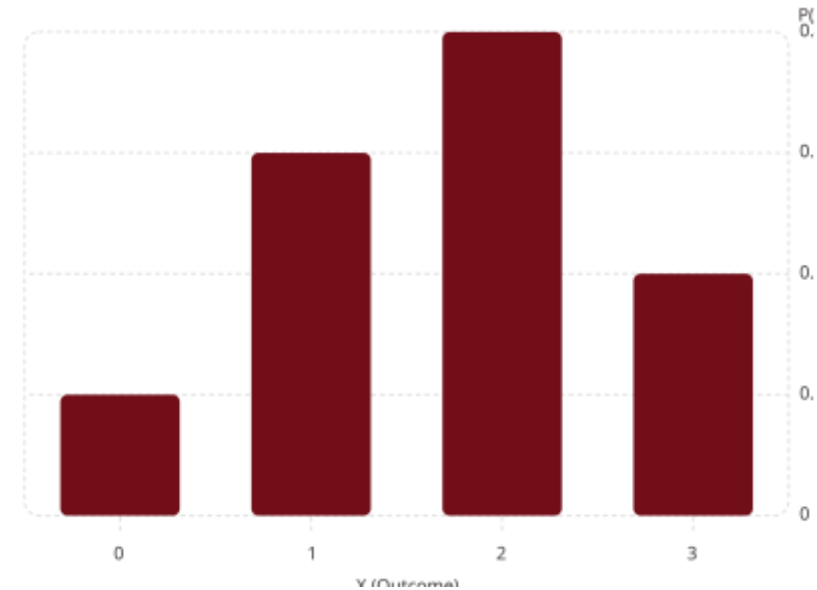


Discrete Probability Distribution

A **discrete distribution** deals with **countable outcomes**. The Probability Mass Function (PMF) must satisfy:

$$0 \leq P(X = x) \leq 1$$

$$\sum P(X = x) = 1$$



☐ Coin Toss

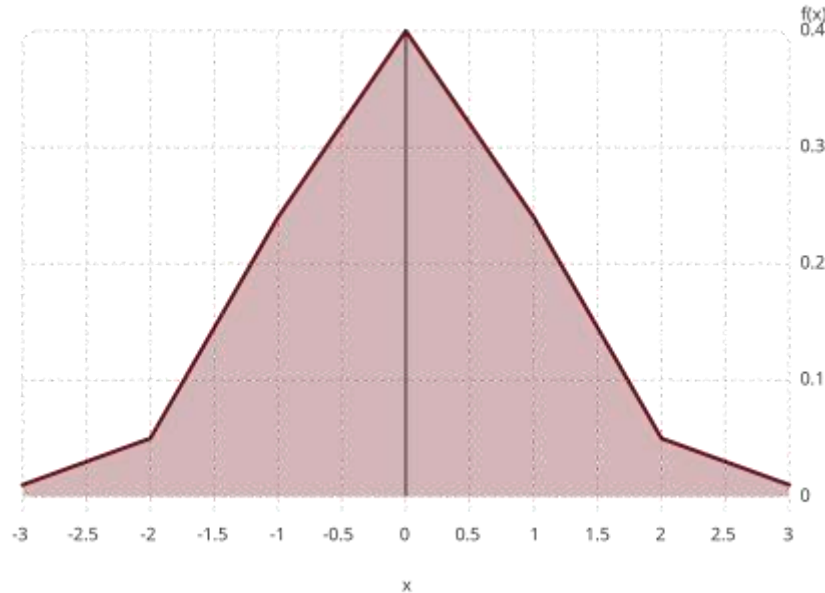
Number of heads in 3 flips

💡 Defective Bulbs

Defects in a batch of 10

Continuous Probability Distribution

A **continuous distribution** handles measurable values over a range. Probability is the **area under the curve** between two points.



$$P(a < X < b) = \int_a^b f(x) dx$$

- $f(x) \geq 0$ for all x
- Total area = 1

Examples: student height, temperature, body weight

Mean / Expected Value

Discrete Formula

$$\mu = \sum x \cdot P(x)$$

Continuous Formula

$$\mu = \int x f(x) dx$$

Solved Example

Find the expected value when X takes values 1, 2, 3 with probabilities 0.2, 0.5, 0.3:

$$\mu = 1(0.2) + 2(0.5) + 3(0.3) = 2.1$$

The mean $\mu = 2.1$ is the long-run average — the balance point of the distribution.

Variance and Standard Deviation

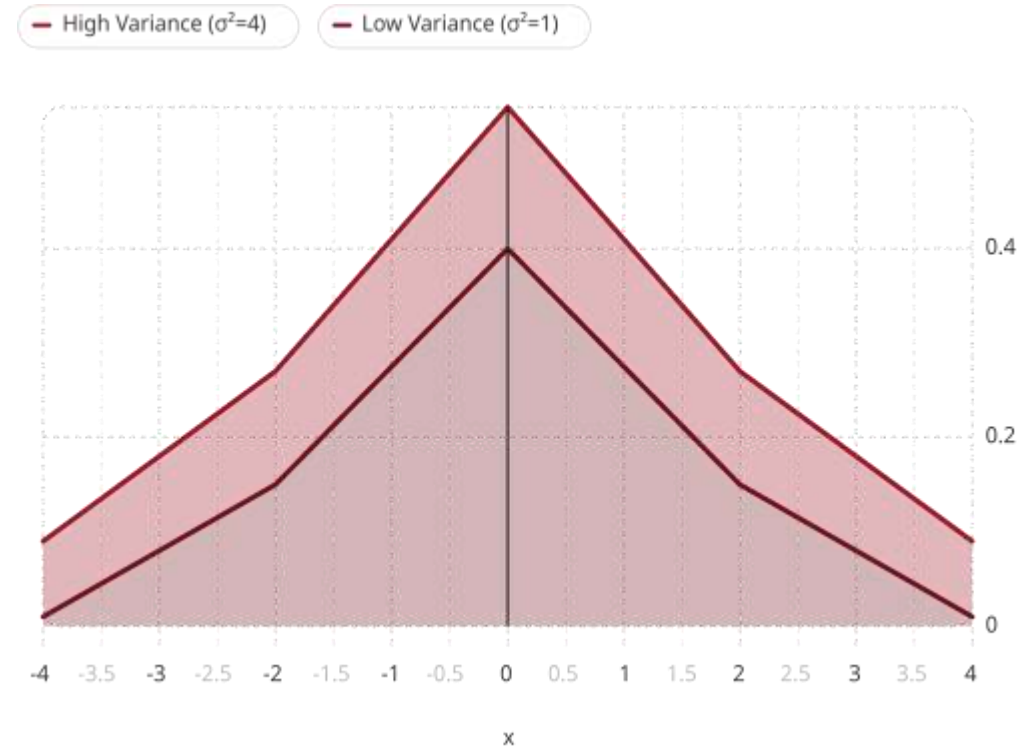
Formulas

$$\sigma^2 = \sum (x - \mu)^2 P(x)$$

$$\sigma = \sqrt{\sigma^2}$$

Small variance → values clustered near the mean.

Large variance → data spread widely.



Binomial Distribution

Models the number of **successes** in **n fixed, independent trials**, each with constant probability **p**.

Formula

$$P(X = x) = \binom{n}{x} p^x (1 - p)^{n-x}$$

Mean & Variance

$$\mu = np$$

$$\sigma^2 = np(1 - p)$$



Example: $P(\text{exactly 3 heads in 5 tosses}) = C(5,3) \cdot (0.5)^3 \cdot (0.5)^2 \approx 0.3125$

Poisson Distribution

Models the number of **events** occurring in a fixed interval at a constant average rate λ .

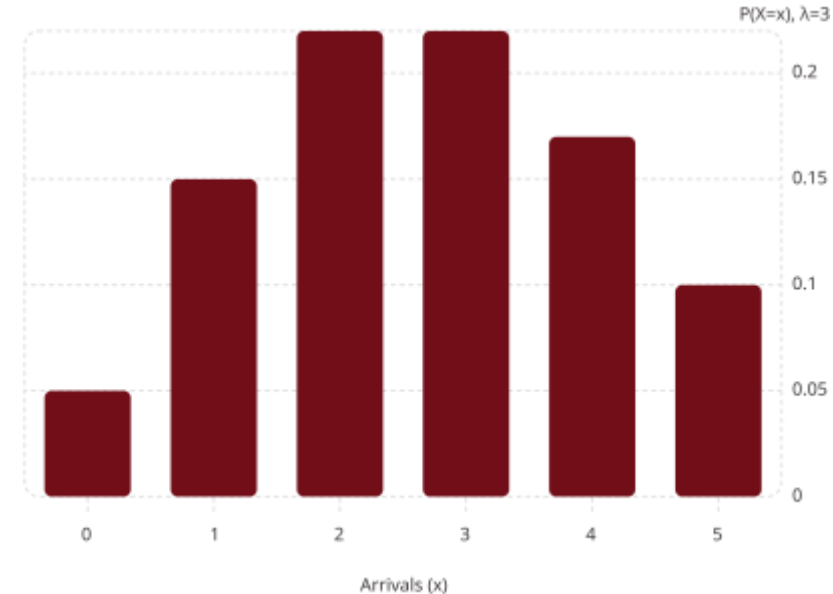
$$P(X = x) = \frac{e^{-\lambda} \lambda^x}{x!}$$

- Mean: $\mu = \lambda$
- Variance: $\sigma^2 = \lambda$



Example: $\lambda = 3$ customers/hour.

$$P(X = 2) = e^{-3} \cdot 3^2 / 2! \approx 0.224$$



Binomial vs. Poisson: Comparison

Both model discrete outcomes but apply to different scenarios.

Feature	□ Binomial	□ Poisson
Trial Type	Fixed number of trials (n)	Fixed time or space interval
Outcomes	Success / Failure	Count of events
Parameters	n (trials), p (probability)	λ (average rate)
Mean	$\mu = np$	$\mu = \lambda$
Variance	$\sigma^2 = np(1 - p)$	$\sigma^2 = \lambda$
Example	3 heads in 5 coin tosses	2 customers arriving per hour

KEY TAKEAWAYS

Remember



Random Variables

Can be **discrete** (countable)
or **continuous** (measurable)



Mean (μ)

Measures the **center** or long-run
average of a distribution



Variance (σ^2)

Measures the **spread** of
values around the mean



Binomial

Used for **fixed trials** with two outcomes



Poisson

Used for **events over time or space** at a constant rate

Thank You..